Taylor Tables of Differencing Schemes

- 1. Notation: Consider u(x,t) for fixed t and $x=j\Delta x$ so that, $u(x+k\Delta x)=u(j\Delta x+k\Delta x)=u_{j+k}$.
- 2. The generalized form of the Taylor Series Expansions is given by

$$u_{j+k} = u_j + (k\Delta x) \left(\frac{\partial u}{\partial x}\right)_j + \frac{1}{2} (k\Delta x)^2 \left(\frac{\partial^2 u}{\partial x^2}\right)_j + \dots + \frac{1}{n!} (k\Delta x)^n \left(\frac{\partial^n u}{\partial x^n}\right)_j + \dots$$

3. For example, consider the Taylor series expansion for u_{j+1} :

$$u_{j+1} = u_j + (\Delta x) \left(\frac{\partial u}{\partial x}\right)_j + \frac{1}{2} (\Delta x)^2 \left(\frac{\partial^2 u}{\partial x^2}\right)_j + \dots + \frac{1}{n!} (\Delta x)^n \left(\frac{\partial^n u}{\partial x^n}\right)_j + \dots$$

4. Or for u_{j-2} :

$$u_{j-2} = u_j + (-2\Delta x) \left(\frac{\partial u}{\partial x}\right)_j + \frac{1}{2} (-2\Delta x)^2 \left(\frac{\partial^2 u}{\partial x^2}\right)_j + \dots + \frac{1}{n!} (-2\Delta x)^n \left(\frac{\partial^n u}{\partial x^n}\right)_j + \dots$$

Taylor Table For the 1st Order Backward Difference

1. Given

$$\left(\frac{\partial u}{\partial x}\right)_{i} - \frac{(u_{j} - u_{j-1})}{\Delta x} = er_{t}$$

- 2. Each term is expanded in it's Taylor Series and placed in a table to simplify the algebra.
- 3. Note the multiplication by Δx to again simplify the table.

$$u_{j} \qquad \begin{pmatrix} \Delta x \cdot & \Delta x^{2} \cdot & \Delta x^{3} \cdot & \Delta x^{4} \cdot \\ \left(\frac{\partial u}{\partial x}\right)_{j} & \left(\frac{\partial^{2} u}{\partial x^{2}}\right)_{j} & \left(\frac{\partial^{3} u}{\partial x^{3}}\right)_{j} & \left(\frac{\partial^{4} u}{\partial x^{4}}\right)_{j} \\ & - & - & - & - & - \\ \Delta x \cdot \left(\frac{\partial u}{\partial x}\right)_{j} & 1 & 1 & (-1)\frac{1}{1!} & (-1)^{2}\frac{1}{2!} & (-1)^{3}\frac{1}{3!} & (-1)^{4}\frac{1}{4!} \\ & - u_{j} & -1 & 0 & 0 & 0 & 0 \\ & = & - & - & - & - \\ \Delta x \cdot er_{t} & 0 & 0 & \frac{1}{2} & ? & ? \end{cases}$$

- 4. The truncation error term $er_t = \frac{1}{2}\Delta x \left(\frac{\partial^2 u}{\partial x^2}\right)_j$ is defined from the first non-zero column.
- 5. Don't forget the division by the Δx to undo the previous multiplication.
- 6. Order of accuracy is defined as the exponent on the Δx term in er_t .

Taylor Table For the 2^{nd} Order Central Difference

1. Given

$$\left(\frac{\partial u}{\partial x}\right)_{j} - \frac{(u_{j+1} - u_{j-1})}{2\Delta x} = er_{t}$$

2. The Taylor Table

- 3. The truncation error term $er_t = -\frac{1}{6}\Delta x^2 \left(\frac{\partial^3 u}{\partial x^3}\right)_j$ is defined from the first non-zero column.
- 4. Accuracy is 2^{nd} Order.

Taylor Table For A General 3 Point Difference Scheme

1. Starting with

$$\left(\frac{\partial u}{\partial x}\right)_j - \frac{1}{\Delta x}(c u_{j-2} + b u_{j-1} + a u_j) = er_t$$

2. The Taylor Table

3. Now instead of having colums sum to zero, we set enough colums to zero to satisfy the number of unknowns.

Taylor Table For A General 3 Point Difference Scheme

1. This time the first three columns sum to zero if

$$\begin{bmatrix} -1 & -1 & -1 \\ 2 & 1 & 0 \\ -4 & -1 & 0 \end{bmatrix} \begin{bmatrix} c \\ b \\ a \end{bmatrix} = \begin{bmatrix} 0 \\ -1 \\ 0 \end{bmatrix}$$

- 2. Note we put the linear equations into a matrix form, let Matlab do the work for you.
- 3. Which gives $[c, b, a] = \frac{1}{2}[1, -4, 3]$.
- 4. In this case the fourth column provides the leading truncation

$$er_t = \frac{1}{\Delta x} \left[\frac{8c}{6} + \frac{b}{6} \right] \Delta x^3 \left(\frac{\partial^3 u}{\partial x^3} \right)_j = \frac{\Delta x^2}{3} \left(\frac{\partial^3 u}{\partial x^3} \right)_j$$

5. Thus we have derived a second-order backward-difference approximation of a first derivative:

$$\left(\frac{\partial u}{\partial x}\right)_j = \frac{1}{2\Delta x}(u_{j-2} - 4u_{j-1} + 3u_j) + O(\Delta x^2)$$

Taylor Table For Other Derivatives, e.g. 2^{nd}

1. Consider a gerneral 3 point formula for the 2^{nd} derivative

$$\left(\frac{\partial^2 u}{\partial x^2}\right)_j - \frac{1}{\Delta x^2} (a u_{j-1} + b u_j + c u_{j+1}) = er_t$$

2. The Taylor Table is

3. Setting the first 3 colums to 0 leads to

$$\begin{bmatrix} -1 & -1 & -1 \\ 1 & 0 & -1 \\ -1 & 0 & -1 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ -2 \end{bmatrix}$$

4. The solution is given by [a, b, c] = [1, -2, 1].

Taylor Table For 2^{nd} Derivative

1. In this case er_t occurs at the fifth column in the table (for this example all even columns will vanish by symmetry) and one finds

$$er_t = \frac{1}{\Delta x^2} \left[\frac{-a}{24} + \frac{-c}{24} \right] \Delta x^4 \left(\frac{\partial^4 u}{\partial x^4} \right)_j = \frac{-\Delta x^2}{12} \left(\frac{\partial^4 u}{\partial x^4} \right)_j$$

- 2. Note that Δx^2 has been divided through to make the error term consistent.
- 3. We have just derived the familiar 3-point central-differencing point operator for a second derivative

$$\left(\frac{\partial^2 u}{\partial x^2}\right)_j = \frac{1}{\Delta x^2}(u_{j-1} - 2u_j + u_{j+1}) + O(\Delta x^2)$$